



Model Calibration Methods for Phase Diagram Determination

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Coffee: Plain black coffee, brewed less than an hour ago.





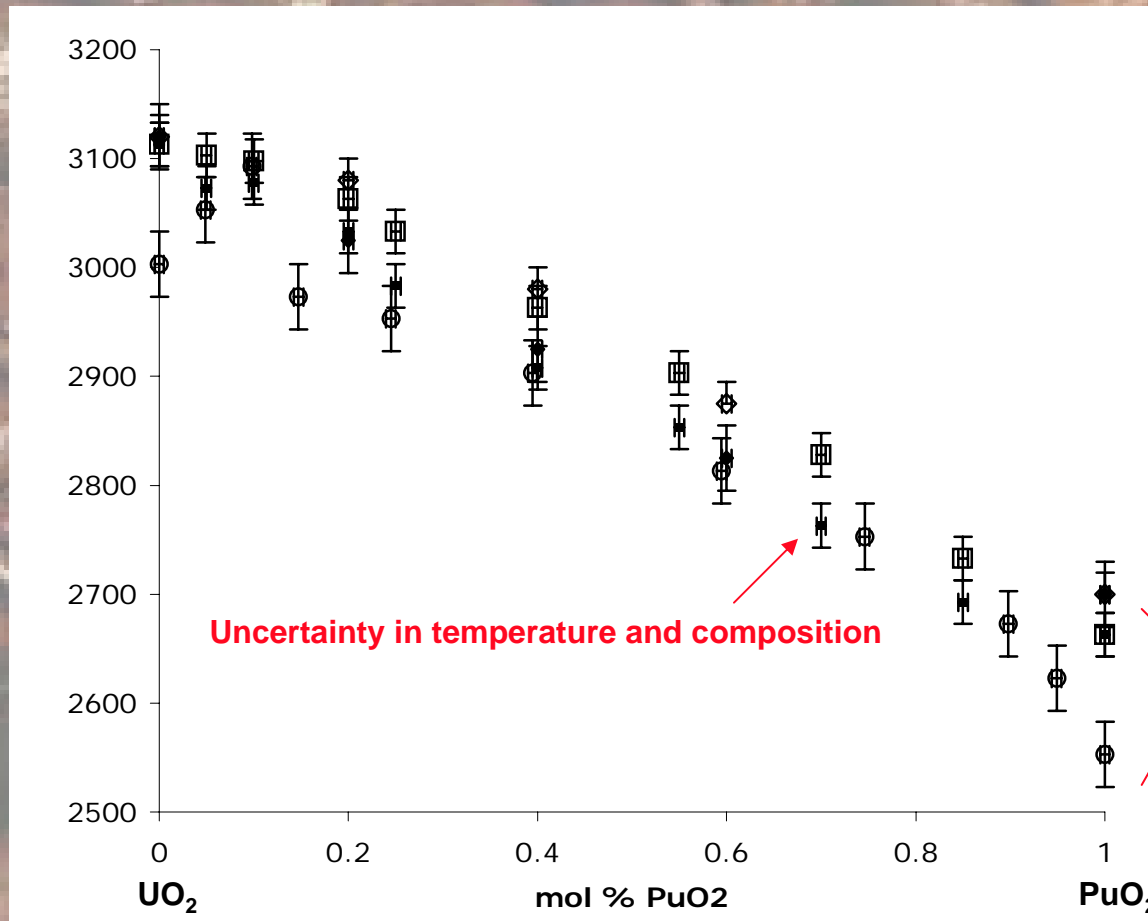
Why Model Phase Diagrams

- 5 to 10 component phase systems are often used in critical applications
 - Nuclear fuels
- One must predict phase transitions in these multi-component systems
 - melting points and eutectic compositions
 - volume changes
- However, it is not feasible to experimentally determine the entire phase diagram of a multi-component system
- Thus the need for modeling.



Uncertainty In Phase Diagrams

Where are
the solidus
and
liquidus?





The Uncertainty of the $\text{UO}_2\text{-PuO}_2$ Phase Diagram*

Collect available data



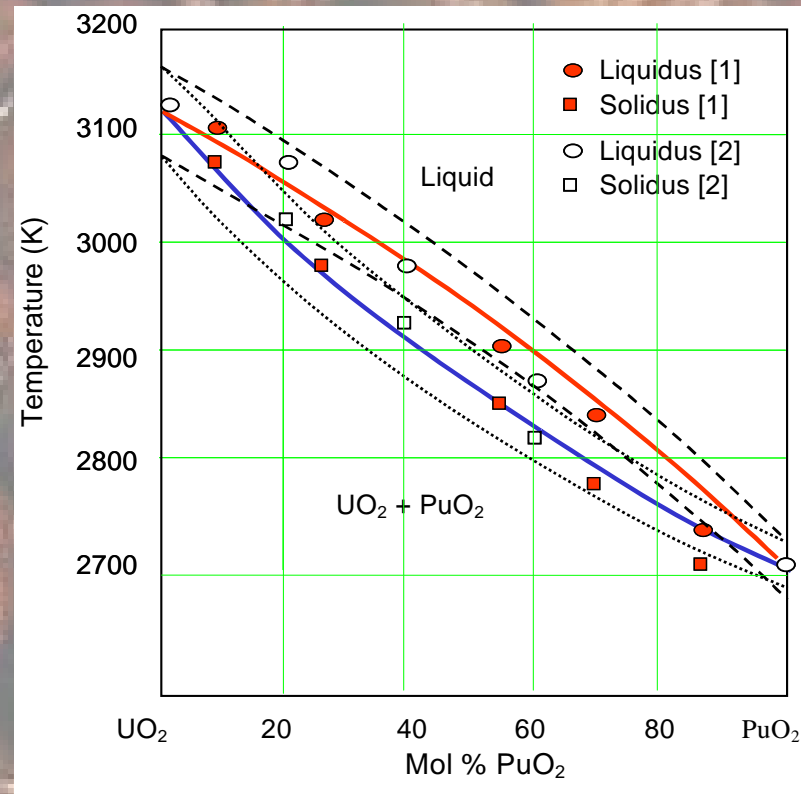
Bayesian Statistics
and Genetic Algorithm



Evaluate uncertainty



Calculate diagram



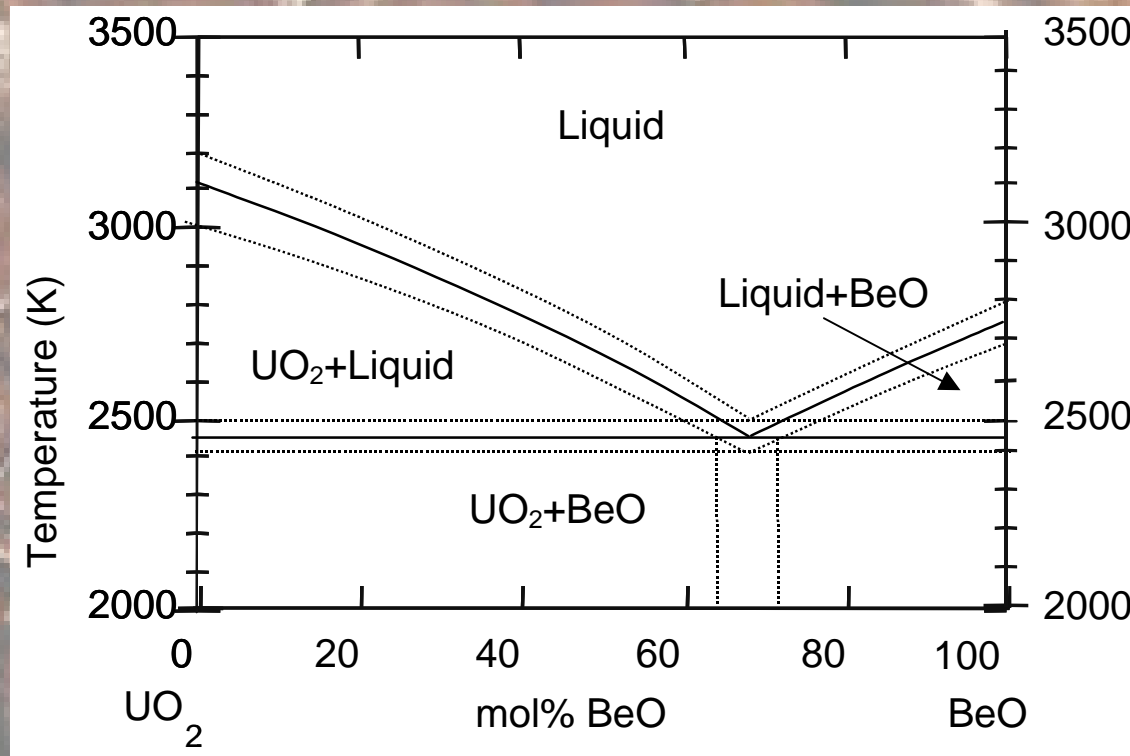
*M. Stan and B. J. Reardon, *CALPHAD*, 27 (2003) 319-323.

[1] M. G. Adamson, E. A. Aitken, and R. W. Caputi, *J. Nucl. Mater.*, 130 (1985) 349-365.

[2] T. D. Chikalla, *J. Am. Ceram. Soc.*, 47 (1964) 309-309.



The Uncertainty of the $\text{UO}_2\text{-BeO}$ Phase Diagram*



*M. Stan and B. J. Reardon, *CALPHAD*, 27 (2003) 319-323.

Experimental by P. P. Budnikov, S. G. Tresvyatski, and V. I. Kushakovskiy, Proc. 2nd U. N. Intern. Conf. Peaceful Uses At. Energy, Geneva, 1958, pp. 127.



Sources of Uncertainty

- Uncertainty in phase boundaries are due to:
 - Difficulty in measuring temperature
 - Difficulty in identifying the onset of a phase transition
 - Composition drift due to vaporization
- Many thermodynamic values of components are also uncertain
 - Melting Temperature (T^M)
 - Heat of Melting (ΔH^M)
- For these reasons, many authors report vastly different values for boundary positions



Which model should you use?

1) Ideal Solid Solution Law

$$x^{Liq}(T) = \frac{1 - \exp\left(\frac{\Delta H_{UO_2}^M}{R} \left(\frac{1}{T} - \frac{1}{T_{UO_2}^M}\right)\right)}{\exp\left(\frac{\Delta H_{PuO_2}^M}{R} \left(\frac{1}{T} - \frac{1}{T_{PuO_2}^M}\right)\right) - \exp\left(\frac{\Delta H_{UO_2}^M}{R} \left(\frac{1}{T} - \frac{1}{T_{UO_2}^M}\right)\right)}$$

$$x^{Sol}(T) = x^{Liq}(T) \cdot \exp\left(\frac{\Delta H_{PuO_2}^M}{R} \left(\frac{1}{T} - \frac{1}{T_{PuO_2}^M}\right)\right)$$

2) Polynomial in X

$$T_s(K) = a_s + b_s x + c_s x^2 + d_s x^3$$

$$T_l(K) = a_l + b_l x + c_l x^2$$

5) Model 1 & the eutectic UO_2 -BeO system

$$x_{Liq+BeO}^{Liq}(T) = \exp\left(\left(\frac{-\Delta H_{BeO}^M}{RT}\right) \ln\left(\frac{T_{BeO}^M}{T}\right)\right)$$

$$x_{UO_2+Liq}^{Liq}(T) = 1 - \exp\left(\left(\frac{-\Delta H_{UO_2}^M}{RT}\right) \ln\left(\frac{T_{UO_2}^M}{T}\right)\right)$$

3) Polynomial in X

$$T_s(K) = a_s + b_s x + c_s x^2$$

$$T_l(K) = a_l + b_l x + c_l x^2$$

4) Polynomial in X

$$T_s(K) = T_{MUO_2} / (1 + b_s x + c_s x^2)$$

$$T_l(K) = T_{MUO_2} / (1 + b_l x + c_l x^2)$$

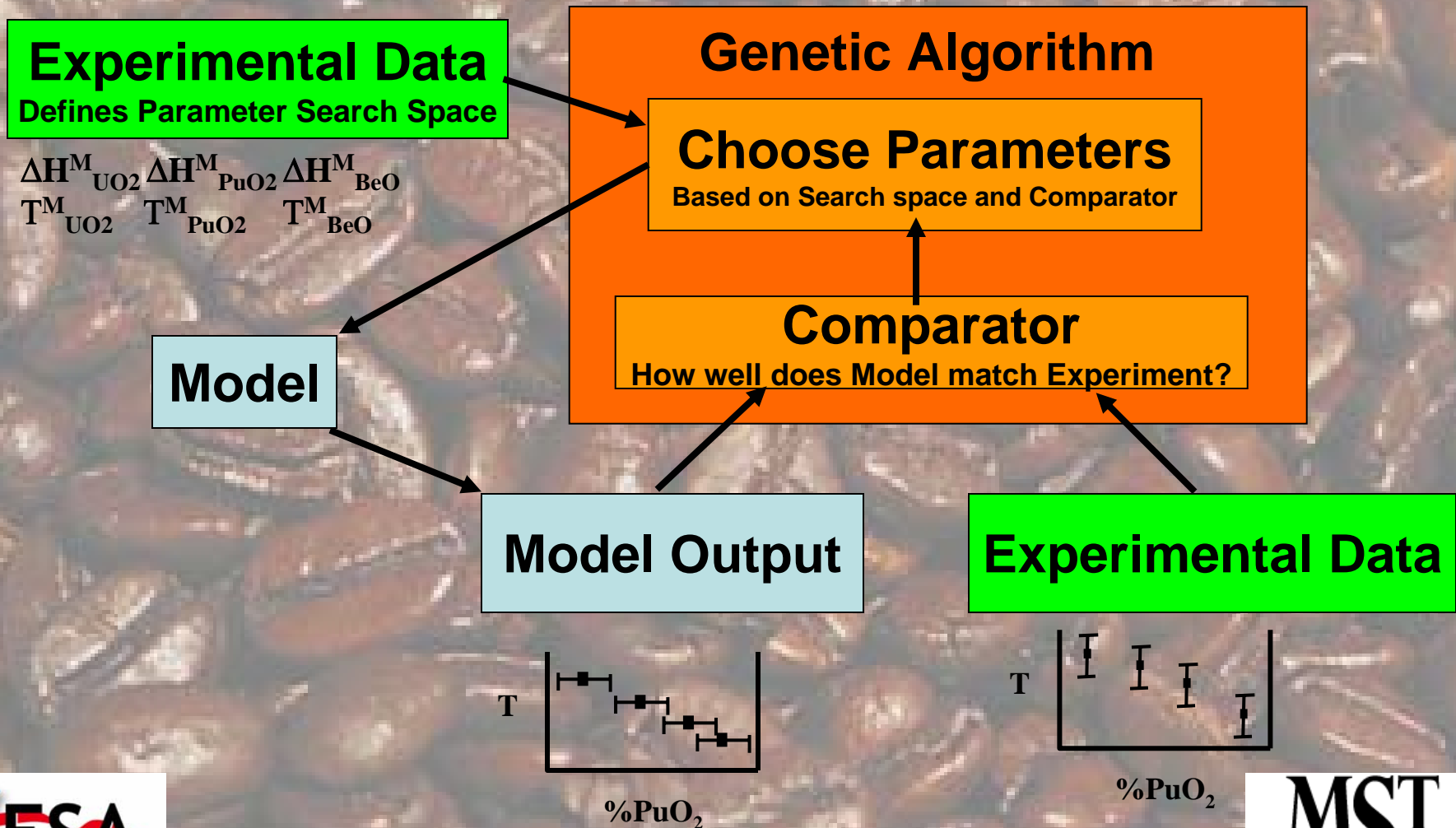


Pros and Cons of each Model

Mode I	Pros	Cons
1	Based on thermodynamics Model parameters applicable to other systems Model parameters experimentally accessible	Assumes ideal solution Model parameters uncertain Model: $x(T)$, Data: $T(x)$ - hard to fit
2,3,4	Model and data : $T(x)$ - much easier to fit	Arbitrary functions and parameters not applicable to other systems Parameters not experimentally accessible
5	Same as Model 1 Incorporating another phase diagram constrains $\Delta H_{UO_2}^M$ and $T_{UO_2}^M$	Same as Model 1 Requires ΔH_{BeO}^M T_{BeO}^M Assumes perfect eutectic.



The Calibration Problem





Why Use a Genetic Algorithm?

- Robust to many classes of problems
- Does not assume distributional form of uncertainties
- Provides distributions and correlations of parameter values
- Using fuzzy rule set, the GA compares any number of experimental data points to model results
- The distribution of optimal solutions provides insight to experimental design*.

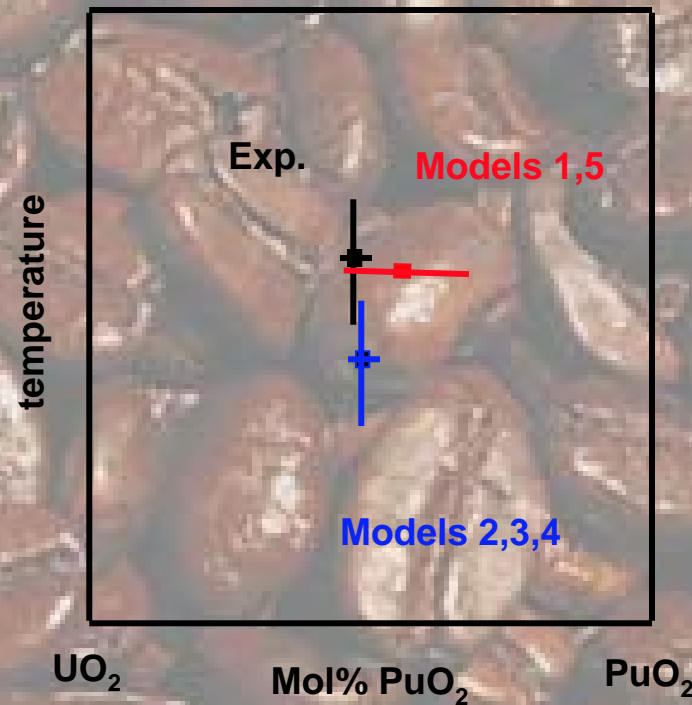
B.J. Reardon, S. Bingert, *Acta Materialia*, 2000, 48(3), p.647-58



How to compare $x(T)$ with $T(x)$?

Easy to compare:

- RMS
- χ^2
- Kolmogorov-Smirnov
- Kullback-Liebler
- Jeffery's J
- Fuzzy rule set

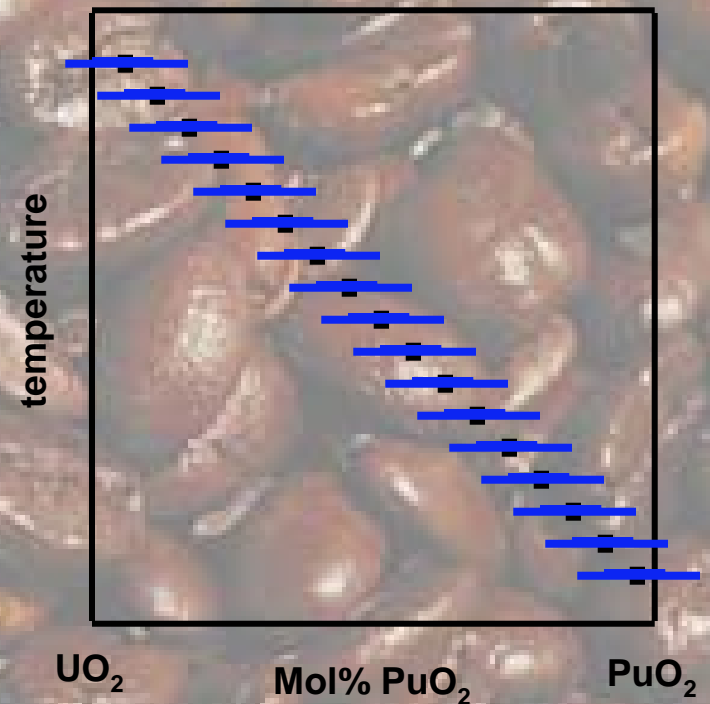
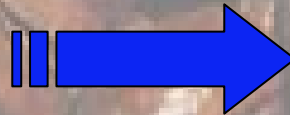
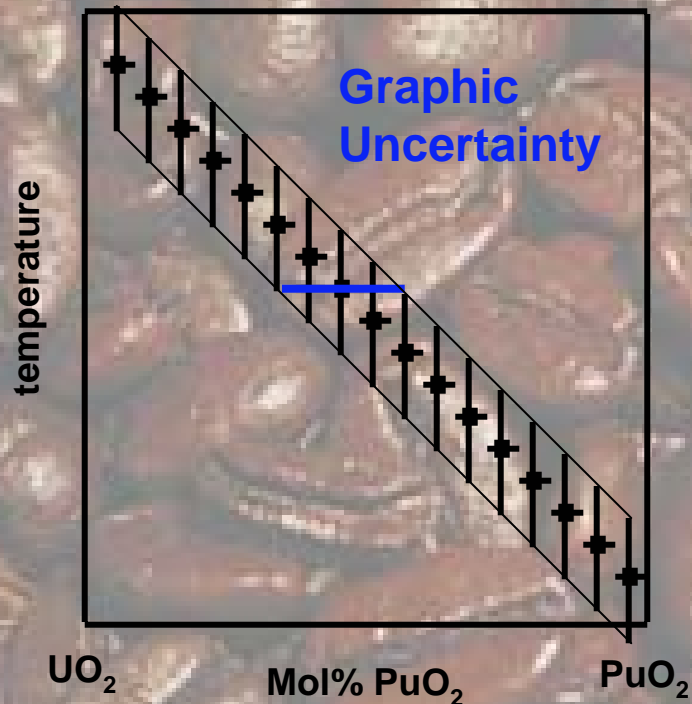


Hard to compare:

•?



Converting uncertainty in $T(x)$ to $x(T)$





Applying the GA to this Calibration Problem

- Define model to be studied
- Define search range for model parameters
- Run GA
- Analyze the number and fitness of solutions
- Analyze how well each model was fit by the GA



Results of Model 1 GA Calibration

The results of optimizing model 1 against the available data sets.

C: Chikalla, L: Lyon and Baily, and A: Aitken and Evans.

	Model	Data Sets	# Sol	Fitness
1	1a	L	1	0.949274
2	1a	A	1	0.976892
3	1a	C	1	0.84066
4	1a	L+A	1	0.928932
5	1a	L+A+C	1	0.790519
6	1b	L	394	1
7	1b	A	1	0.989953
8	1b	C	1	0.887064
9	1b	L+A	1	0.99024
10	1b	L+A+C	1	0.874315

- **Model 1a vs. Model 1b**

- Model 1a: standard uncertainty in exp. X
- Model 1b 'Graphically driven' uncertainty in exp. X

- **# Solutions**

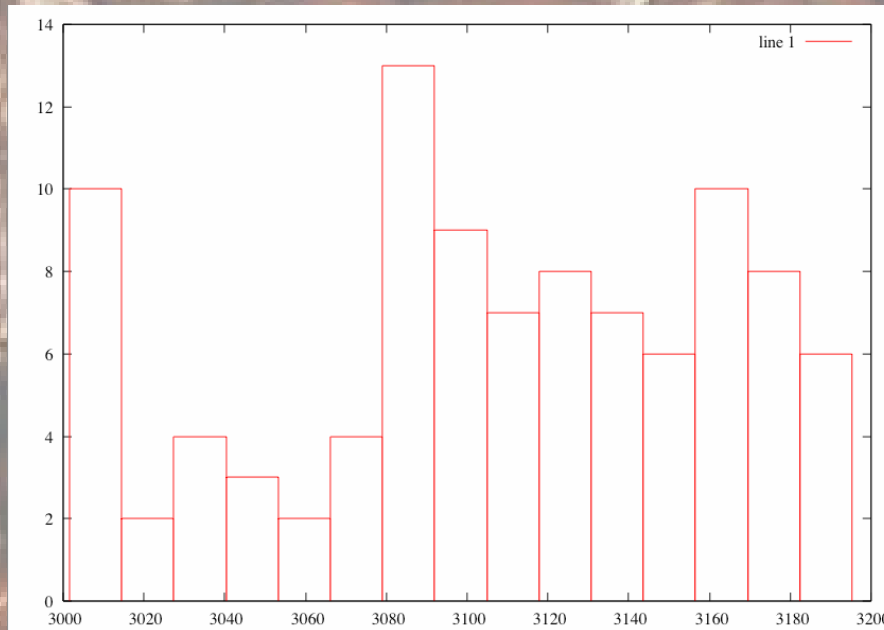
- number of solutions found
- Lyon's data best fits the thermodynamic model

- **Fitness:**

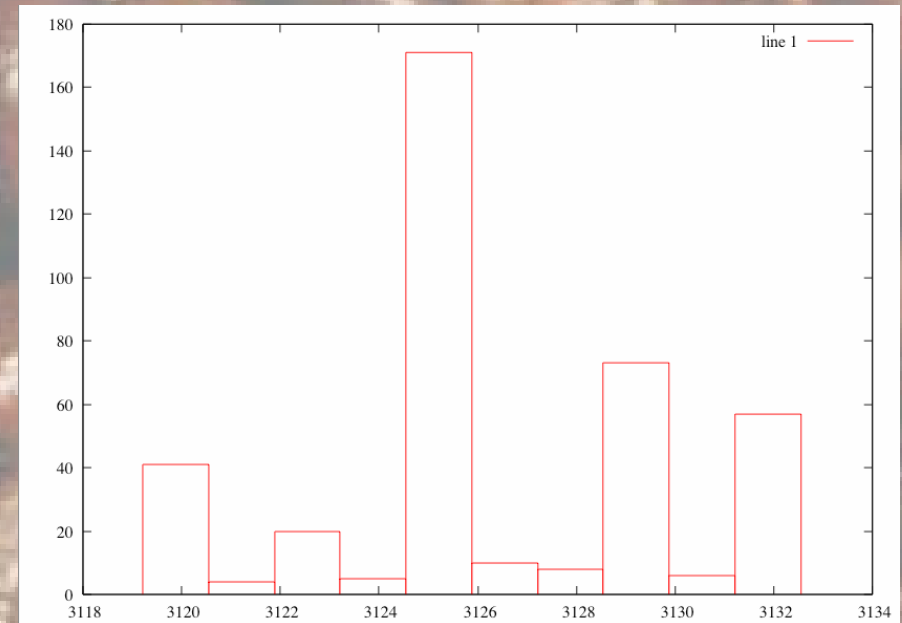
- 1 is the maximum - perfect fit
- Goes up with 'Graphically driven' uncertainty



Evolution of $\text{UO}_2\text{-T}^{\text{M}}$



Initial



**Final - 394 solutions
after 72 generations**



Results of Model 2 GA Calibration

The results of optimizing model 2 against the available data sets.

C: Chikalla, L: Lyon and Baily, and A: Aitken and Evans.

	Model	Data Sets	# Solutions	Fitness
11	2	L	5	0.998358
12	2	A	377	1
13	2	C	1	0.9591
14	2	L+A	14	0.995911
15	2	L+A+C	1	0.941159

- **# Solutions**

- This model was originally developed for Aitken's data. This can be seen in the fact that a large number of solutions were found when using only Aitken's data set.
- Unfortunately, this model can not be extended to any other phase system.



Results of Model 3 GA Calibration

The results of optimizing model 3 against the available data sets.

C: Chikalla, L: Lyon and Baily, and A: Aitken and Evans.

	Model	Data Sets	# Solutions	Fitness
16	3	L	145	0.998775
17	3	A	291	1
18	3	C	502	1
19	3	L+A	82	0.993154
20	3	L+A+C	2	0.930555

- **# Solutions**

- This model was originally developed Chikalla's data. This can be seen in the fact that a large number of solutions were found when using only Chikalla's data set.
- Unfortunately, this model can not be extended to any other phase system.
- Also, it should be noted that the optimized parameters from each run (16-20) are significantly different.



Results of Model 4 GA Calibration

The results of optimizing model 4 against the available data sets.

C: Chikalla, L: Lyon and Baily, and A: Aitken and Evans.

	Model	Data Sets	# Solutions	Fitness
21	4	L	3	0.983283
22	4	A	1	0.993808
23	4	C	449	0.960122
24	4	L+A	1	0.982623
25	4	L+A+C	1	0.920988

- **# Solutions**

- This model does not fit any of the data sets well
- Like the others, this model can not be extended to any other phase system.
- The large number of solutions found when using Chikalla's data is not significant since the over all fitness of these solutions is so low.

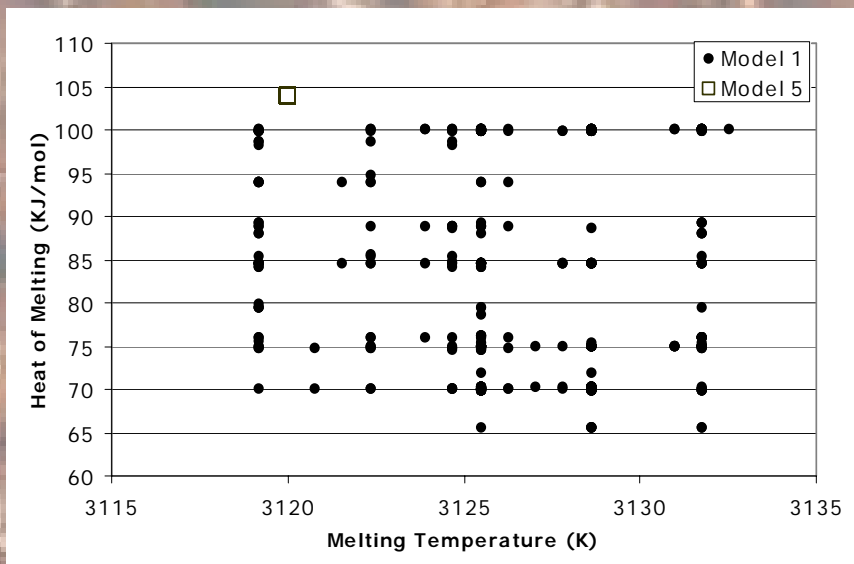


Results of Model 5 GA Calibration

The results of optimizing model 5 against the available data sets.

C: Chikalla, L: Lyon and Baily, and A: Aitken and Evans.

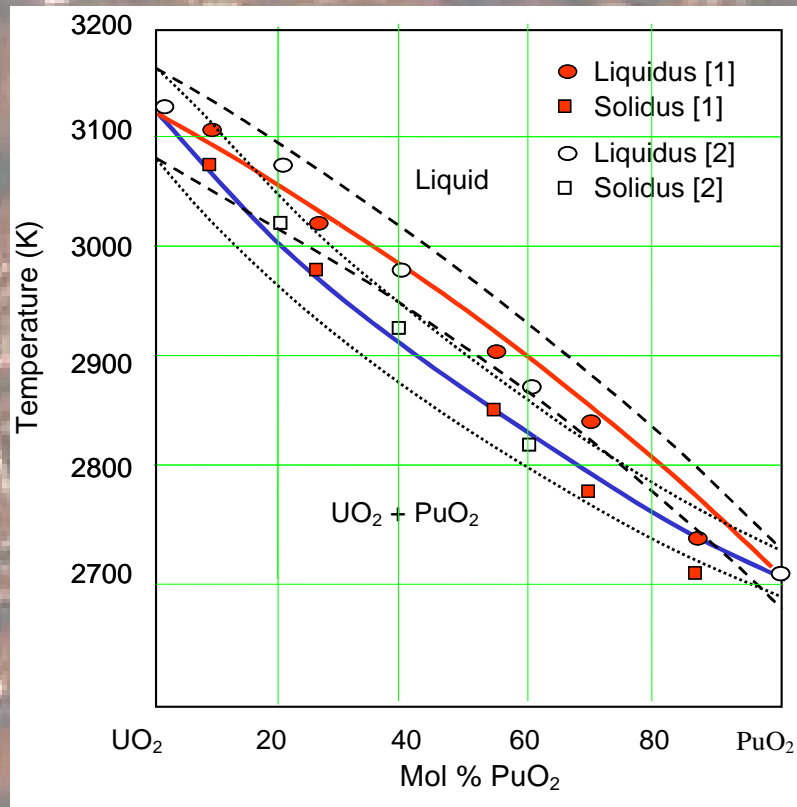
Test	Model	Data Sets	# Solutions	Fitness
26	5a	L	11	0.974211
27	5a	A	322	0.981022
28	5a	C	105	0.899736
29	5a	L+A	1	0.963978
30	5a	L+A+C	7	0.894648
31	5b	L	308	0.999815
32	5b	A	255	1
33	5b	C	395	0.933822
34	5b	L+A	43	0.995135
35	5b	L+A+C	127	0.930271



The final solution sets for the heats of melting and the melting points of UO_2 determined through the optimization of Model 1 (circle) and Model 5 (square).



The Uncertainty of the $\text{UO}_2\text{-PuO}_2$ Phase Diagram*



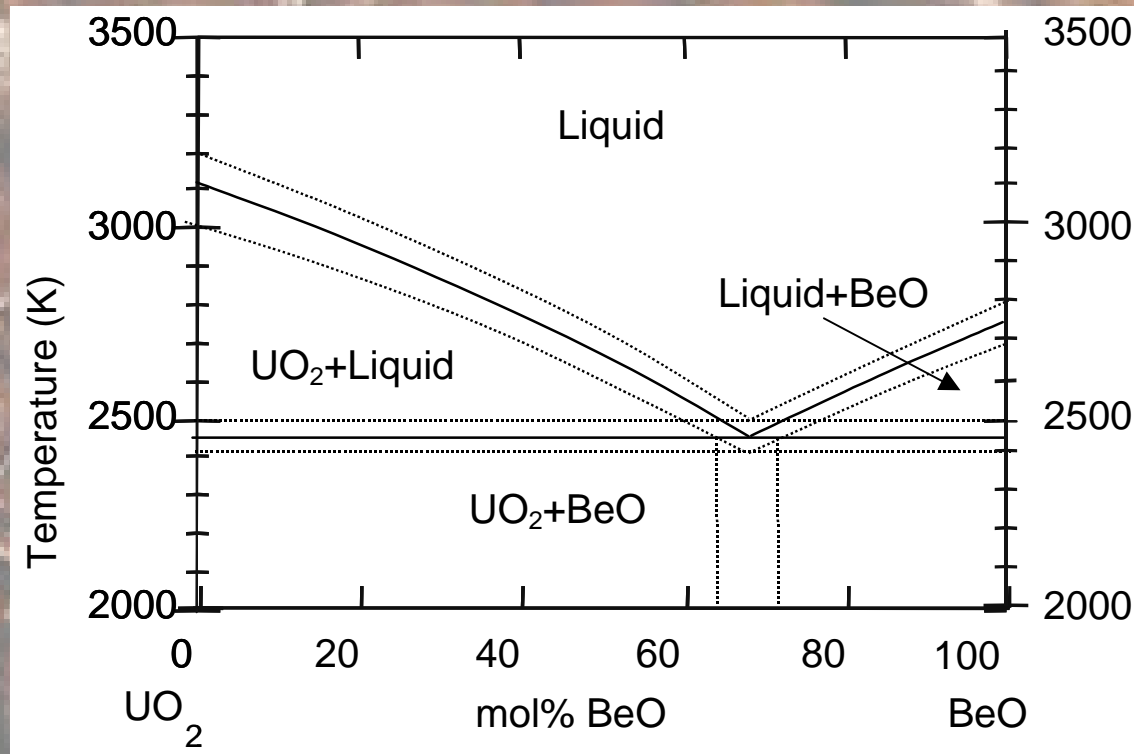
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The Uncertainty of the UO_2 -BeO Phase Diagram*



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Experimental by P. P. Budnikov, S. G. Tresvyatski, and V. I. Kushakovsky, Proc. 2nd U. N. Intern. Conf. Peaceful Uses At. Energy, Geneva, 1958, pp. 127.



Summary of Calibration Results

- The ‘graphic uncertainty’ conversion was necessary to find a large number of solutions given the thermodynamic model
- While the polynomial models fit some of the data sets well they
 - Do not handle all the data well
 - Can not be extended to other systems
- Some data sets are more thermodynamically self consistent than others



Conclusions

- GAs calibrate models using disparate, sparse, uncertain data sources.
- This calibration provides the overall predictive credibility of the models.
- The phase boundary uncertainties of the $\text{UO}_2\text{-PuO}_2$ and $\text{UO}_2\text{-BeO}$ systems have been determined by accounting for:
 - the available phase boundary data
 - the accepted models of the phase boundaries
 - the thermodynamic data used in the models.
- The net result is an internally self-consistent reduction in uncertainty of the values of the thermodynamic data as well as the phase boundaries.
- Modern heuristic optimizers such as GAs were crucial to this work since they are both robust and require no assumptions about the uncertainty distributions.